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GENERATION OF RANDOM VECTORS WITH
KNOWN MEAN AND COVARIANCE

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Fred M. Speed

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NATIONAL AERONAUTICS AND SPACE ADMINISTRATION
MANNED SPACECRAFT CENTER
HOUSTON, TEXAS

June 8, 1965

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GENERATION OF RANDOM VECTORS WITH
KNOWN MEAN AND COVARIANCE

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SUMMARY

In simulation studies, it is necessary to introduce normally distributed random "errors" to the data in order to simulate the actual conditions. These random "errors" are normally distributed random vectors with known mean and covariance. The algorithms currently being used (ref. 3) to generate these vectors encounter one of the two following problems.

- (1) The algorithm can not handle singular covariance matrices (a singular covariance matrix occurs when the data is completely correlated).
- (2) The algorithm requires the inversion of a matrix.

The purpose of this paper is to present an algorithm that does not require the inversion of a matrix and can handle both singular and non-singular covariance matrices. The proposed algorithm should be more efficient than the current algorithms.

INTRODUCTION

This paper is divided into four parts. The first part gives some definitions and theorems that will be needed in the development of the algorithm. The second part contains the development of the algorithm. The third part describes a computer program that generates the desired random vectors. The last part contains Appendices A and B. Appendix A contains a listing of the computer program GERN and presents several examples illustrating the use of the program. Appendix B contains listings of the subroutines that make up the $N(0, 1)$ random number generator packet.

SYMBOLS

X	column vector
Y	column vector
R	positive semi-definite symmetric matrix
Other capital letters	matrices, unless otherwise stated
A^T	transpose of A
A^{-1}	inverse of A
Y is distributed according to $N(\mu, R)$	Y is from a normal population with mean μ and covariance matrix R

$E(Y)$	expected value of Y
$COV(Y)$	covariance of Y
I	identity matrix
ϕ	null vector
D	diagonal matrix

DEFINITIONS AND THEOREMS

Definition 1. Two $n \times n$ real matrices A and B are said to be congruent if there exists a non-singular matrix P such that

$$PAP^T = B.$$

Since congruence is a special case of equivalence, the matrix P can be obtained from elementary row-column operations (ref. 1).

Theorem I. (ref. 1) Every $n \times n$ real symmetric matrix A of rank r is congruent to a diagonal matrix, whose diagonal elements consist of r ones and $n-r$ zeros.

Theorem II. (ref. 2) Let X be distributed according to $N(\phi, H)$. If $Y = AX$, then Y is distributed according to $N(\phi, AHA^T)$.

Theorem III. (ref. 2) If $Y = AX + \mu$, where A is a constant matrix and μ is a constant column vector, then

$$E(Y) = AE(X) + \mu$$

$$COV(Y) = COV(AX)$$

The following theorem is an immediate consequence of matrix theory.

Theorem IV. If D is a diagonal matrix, whose elements consists of zeros and ones, then $D = D \cdot D$.

THE DEVELOPMENT OF THE ALGORITHM

Suppose it is necessary to generate an $n \times 1$ random vector Y with mean ϕ and covariance R . The following theorem provides a new method to obtain this Y .

Theorem V. Let X , an $n \times 1$ vector, be distributed according to $N(\phi, I)$. If R is a $n \times n$ given covariance matrix, then there exists an $n \times n$ matrix A , such that if $Y = AX$, then Y is distributed according to $N(\phi, R)$.

Proof: Let the rank of R be $r \leq n$. By definition, R is a real symmetric matrix. Thus, by Theorem 1, there exists a non-singular matrix P such that $PRP^T = D$, where D is a

diagonal matrix whose diagonal elements consists of r ones and $n-r$ zeros.

Theorem II implies that a necessary and sufficient condition that $Y = AX$ be $N(\phi, R)$ is that $AA^T = R$. It will be shown that $A = RP^T$ is one matrix such that $AA^T = R$.

Since $PRP^T = D$, then

$$(i) \quad RP^T = P^{-1} D$$

$$(ii) \quad PR = D (P^T)^{-1}$$

$$(iii) \quad R = P^{-1} D (P^T)^{-1}$$

Thus, if $A = RP^T$, then

$$\begin{aligned} AA^T &= RP^T PR^T \\ &= RP^T PR \\ &= P^{-1} D D(P^T)^{-1} \\ &= P^{-1} D(P^T)^{-1} \\ &= R \end{aligned}$$

Thus, if $A = RP^T$, then $Y = AX = RP^T X$ is $N(\phi, R)$ and the theorem is proved.

It has been shown (ref. 1) that P can be found by forming the augmented matrix $[R, I]$ and operating on this matrix

with elementary row-column operations in such a way that R is diagonalized. The result of these operations on $[R, I]$ is $[D, P]$. Since X is distributed according to $N(\phi, I)$, the elements of X are readily obtainable from a $N(0, 1)$ random number generator. Thus, given the covariance matrix R , the random vector Y can be obtained rather easily.

SUBROUTINE GERN

GERN is a FORTRAN IV subroutine used to generate random vectors having a known mean and known covariance matrix. All computations are done in single-precision floating point arithmetic. The method used is described in the previous section.

Calling Sequence

Call GERN (N, R, Y, A)

where:

N	size of R
R	covariance matrix. R is dimensioned R(50, 50)

Y n x 1 random vector from a $N(\phi, R)$
population. Y is dimensioned
Y(50)

A a matrix such that $Y = AX$ is
distributed according to $N(\phi, R)$.
A is dimensioned A(50, 50).

Error Messages

If R is not positive semi-definite, a negative number will
occur on the diagonal of R during the row-column operations.
When this occurs, the following message is printed:
"THE ORIGINAL MATRIX IS NOT POSITIVE SEMI-DEFINITE."

Restrictions

$N \leq 50$

$N(0, 1)$ Random Number Generator

See Appendix B

Method

Given: R

Construct: $[R, I]$

Operate on $[R, I]$ with elementary row-column operations to obtain $[D, P]$.

Construct: $X \sim N(0, \Sigma)$

Compute: $A = RP^T$

Compute: $(Y) = AX$

$Y \sim N(0, \Sigma)$

CONCLUSION

Theorem V provides a very simple method to obtain random vectors that can be used to simulate observations that are highly or even completely correlated as well as observations that are independent.

REFERENCES

1. Ayres Jr., Frank: Theory and Problems of Matrices. New York, Schoreem Publishing Co., pp. 40-42, p. 115.
2. Anderson, T. W.: An Introduction to Multivariate Statistical Analysis. New York, John Waley and Sons, Inc., p. 25.
3. Tapley, B. D.: Odell, P. L.: A Study of Optimum Methods for Determining and Predicting Space Vehicle Trajectories and Control Programs. Contract NAS 9-2619.
4. Brunk, H. D.: An Introduction to Mathematical Statistics. Boston, Gin and Co., pp. 86-90.

APPENDIX A

This appendix contains three examples and a listing of the Subroutine GERN.

Example 1: Suppose it is necessary to simulate a 5 x 1 radar observation vector Y, where

$$Y = \begin{bmatrix} \text{range} \\ \text{azimuth} \\ \text{time} \\ \text{elevation} \\ \text{range-rate} \end{bmatrix}$$

In order to simulate the actual conditions, random "observation errors" must be added to the vector Y. Let these "errors" be from a $N(\phi, R)$ population, where

$$R = \begin{bmatrix} 1.0000 & 0.5576 & 0.4641 & 0.8197 & 0.2333 \\ 0.5576 & 2.0000 & 0.1719 & 0.2516 & 0.2265 \\ 0.4641 & 0.1719 & 3.0000 & 0.0264 & 0.0334 \\ 0.8197 & 0.2516 & 0.0264 & 4.0000 & 0.9608 \\ 0.2333 & 0.2265 & 0.0334 & 0.9608 & 5.0000 \end{bmatrix}$$

from a $N(\phi, R)$ population, where R might be

$$R = \begin{bmatrix} 1.0000 & 0.2248 & 0.0000 & 0.9471 & 0.4625 \\ 0.2248 & 2.0000 & 0.0000 & 0.0865 & 0.6449 \\ 0.0000 & 0.0000 & 0.0000 & 0.0000 & 0.0000 \\ 0.9471 & 0.0865 & 0.0000 & 4.0000 & 0.2663 \\ 0.4625 & 0.6449 & 0.0000 & 0.2663 & 5.0000 \end{bmatrix}$$

The matrix A is computed to be

$$A = \begin{bmatrix} 1.0000 & 0.0000 & 0.0000 & 0.0000 & 0.0000 \\ 0.2248 & 1.3962 & 0.0000 & 0.0000 & 0.0000 \\ 0.0000 & 0.0000 & 0.0000 & 0.0000 & 0.0000 \\ 0.9471 & -0.0905 & 0.0000 & 1.7591 & 0.0000 \\ 0.4625 & 0.3873 & 0.0000 & -0.0777 & 2.1517 \end{bmatrix}$$

and the random "error" vector Z is

$$Z = \begin{bmatrix} 0.5879 \\ 0.2784 \\ 0.0000 \\ 1.7551 \\ -0.5159 \end{bmatrix}$$

Example 3: Suppose it is necessary to simulate a 6×1 observation vector Y , where

$$Y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \\ y_5 \\ y_6 \end{bmatrix}$$

$$\text{and where } y_6 = \sum_{i=1}^5 y_i$$

This is a case where an element of the observation vector is completely correlated with the other elements. Let the "errors" be from a $N(\phi, R)$ population where

$$R = \begin{bmatrix} 2.000 & 0.411 & 1.334 & -0.097 & 1.612 & 5.259 \\ 0.411 & 4.000 & -0.238 & -0.684 & -0.656 & 2.832 \\ 1.334 & -0.238 & 6.000 & -1.590 & 1.024 & 6.530 \\ -0.097 & -0.684 & -1.590 & 8.000 & -1.226 & 4.401 \\ 1.612 & -0.656 & 1.024 & -1.226 & 10.000 & 10.755 \\ 5.259 & 2.832 & 6.530 & 4.401 & 10.755 & 29.779 \end{bmatrix}$$

Note that the last row of R is the sum of the first five rows of R and hence R is a singular covariance matrix.

Subroutine GERN computes the matrix

$$A = \begin{bmatrix} 1.414 & 0.000 & 0.000 & 0.000 & 0.000 & 0.002 \\ 0.290 & 1.978 & 0.000 & 0.000 & 0.000 & 0.002 \\ 0.943 & -0.258 & 2.245 & 0.000 & 0.000 & 0.004 \\ -0.069 & -0.335 & -0.718 & 2.714 & 0.000 & 0.001 \\ 1.114 & -0.498 & -0.080 & -0.506 & 2.861 & 0.000 \\ 3.719 & 0.886 & 1.447 & 2.208 & 2.861 & 0.005 \end{bmatrix}$$

The random "error" vector Z is calculated as before and the result is

$$Z = \begin{bmatrix} -1.581 \\ -0.051 \\ 3.311 \\ -0.496 \\ .688 \\ 1.874 \end{bmatrix}$$

A listing of Subroutine GERN follows.

```

*IRFTC GERN
SUBROUTINE GERN(N,R,Y,H)
C     GERN WAS PROGRAMED BY F. M. SPEED MSC , APRIL, 1965
C     CALLING SEQUENCE
C     CALL GERN(N,R,Y,A)
C     WHERE
C     N IS THE SIZE OF R
C     R IS THE COVARIANCE MATRIX, R IS DIMENSIONED R(50,50)
C     Y IS RANDOM VECTOR FROM A N(0,R) POPULATION . Y IS DIMENSIONED
C     Y(50)
C     A IS A MATRIX SUCH THAT YEAX IS DISTRIBUTED ACCORDING TO N(0,
C     A IS DIMENSIONED A(50,50).
C     DIMENSION Y(50),X(50),A(50,100),R(50,50),P(50,50)
A ,D(50,50),H(50,50)
CALL BEGIN(T)
L = 2*N
L1 = N+1
C  CONSTRUCT (R,1)
DO 1 I= 1,N
DO 1 J=L1,L
K = J-N
A(I,J) = 0.0
A(K,J) = 1.0
1  CONTINUE
DO 33 I=1,N
DO 33 J=1,N
A(I,J) = R(I,J)
33  CONTINUE
N1 = N-1
C  COMPUTE (D,P)
DO 2 I = 1,N1
IF(A(I,I))4,5,4
5  CONTINUE
L4 = I + 1
DO 10 K =L4,N
IF(A(K,I))11,10,11
10  CONTINUE
GO TO 4
11  CONTINUE
DO 12 J=1,L
A(I,J) = A(I,J) + A(K,J)
12  CONTINUE
DO 13 J = 1,N
A(J,I) = A(J,I) + A(J,K)
13  CONTINUE
4  CONTINUE
L2 = I+1
DO 6 K=L2,N
D1 = A(K,I)
DO 6 J= 1,L
A(K,J) = A(K,J) - (D1*A(I,J))/A(I,I)
6  CONTINUE
DO 21 J = L2,N
A(I,J) = 0.0
21  CONTINUE
2  CONTINUE
DO 78 I=1,N

```

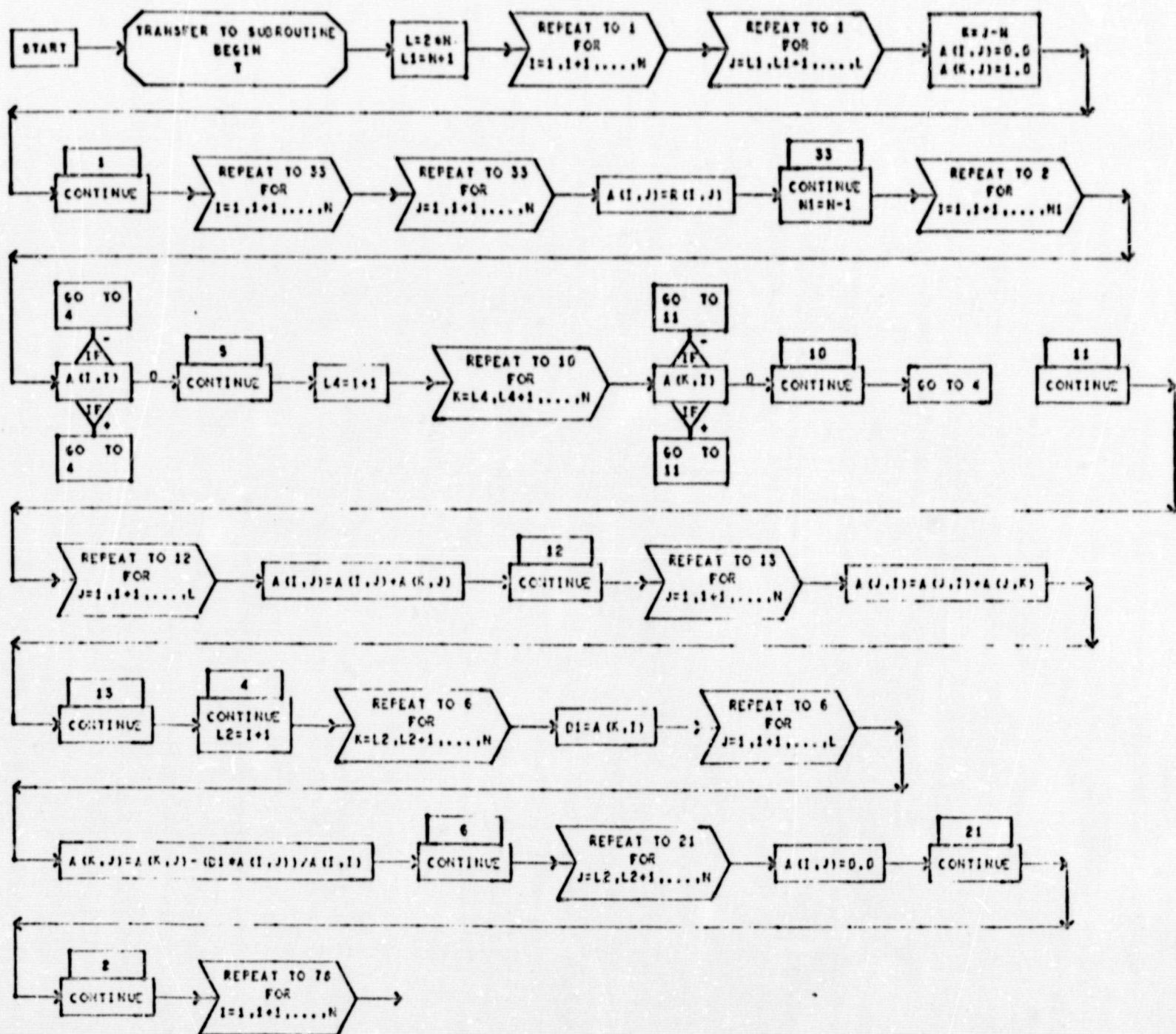


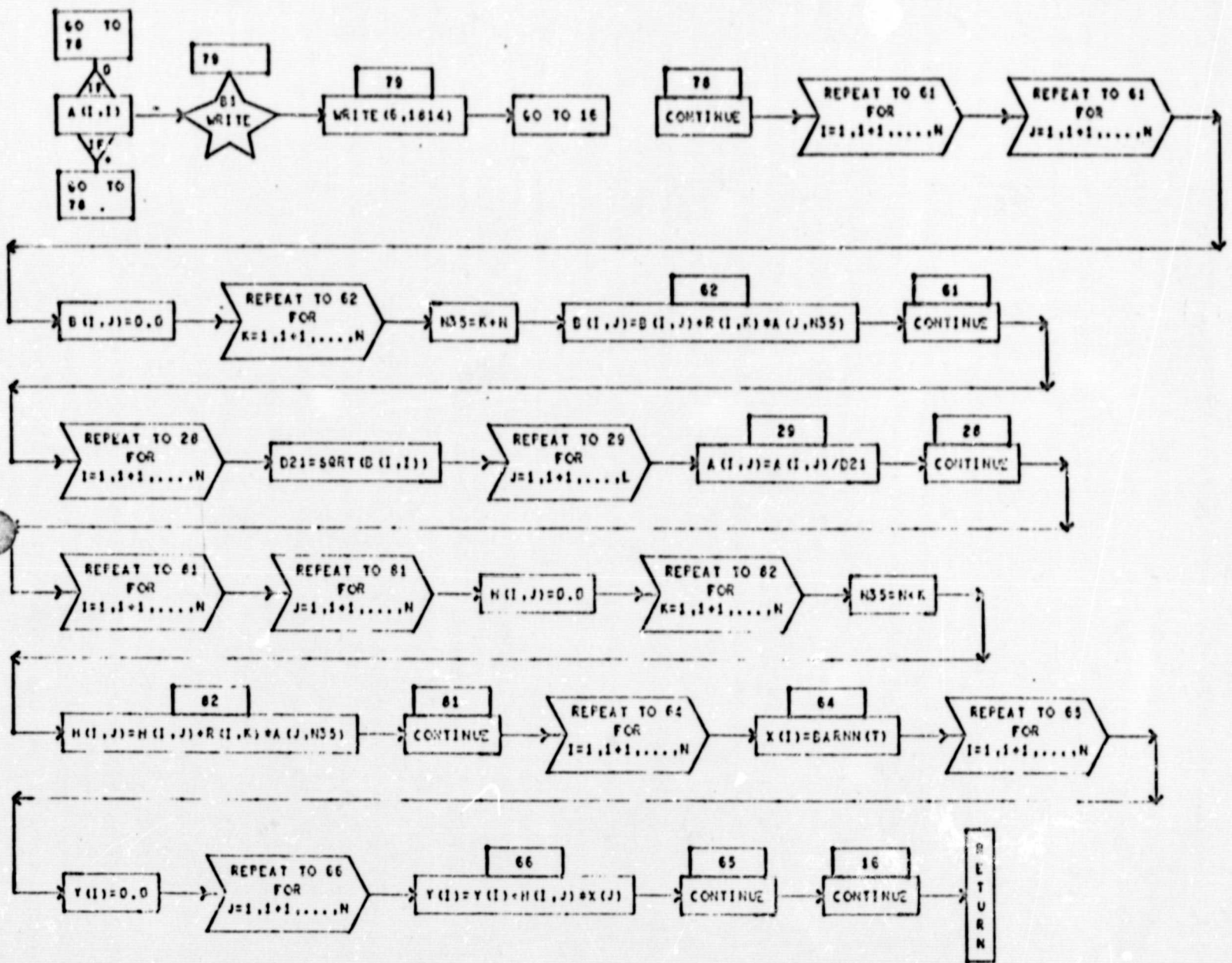
```

      IF(A(I,I)) 79,73,78
79  WRITE(6,1814)
1814  FORMAT(40H THE ORIGINAL MATRIX IS NOT POSITIVE SEMI-DEFINITE)
      GO TO 15
78  CONTINUE
      DO 61 I = 1,N
      DO 61 J = 1,N
      B(I,J) = 0.0
      DO 62 K = 1,N
      N35 = K + N
62  B(I,J) = B(I,J) + R(I,K)*A(J,N35)
61  CONTINUE
      DO 28 I = 1,N
      D21 = SQRT(B(I,I))
      DO 29 J = 1,L
29  A(I,J) = A(I,J)/D21
28  CONTINUE
C  PT = TRANSPOSE OF P
C  COMPUTE RPI
      DO 81 I=1,N
      DO 81 J=1,N
      H(I,J)= 0.0
      DO 82 K=1,N
      N35 = N + K
82  H(I,J) = H(I,J) + R(I,K)*A(J,N35)
81  CONTINUE
C  COMPUTE Y=(RPT)*X
      DO 64 I = 1,N
64  X(I) = RARNV(T)
      DO 65 I = 1,N
      Y(I) = 0.0
      DO 66 J = 1,N
66  Y(I) = Y(I) + H(I,J)*X(J)
65  CONTINUE
16  CONTINUE
      RETURN
      END

```

SUBROUTINE GERN(N,R,Y,H)





APPENDIX B

* N(0, 1) Random Number Generator

The N(0, 1) random number generator packet consist of the following subroutines.

- (i) BEGIN
- (ii) FIXB
- (iii) BARNN
- (iv) BARNA
- (v) LOOKUP
- (vi) BLOCK DATA
- (vii) BARN

All these subroutines are in FORTRAN IV, except BARN, which is in MAP. The first six subroutines in the packet were programmed by K. Oney of Wolf Research and Development Corporation, Houston, Texas. The last subroutine, BARN, is a SHARE subroutine.

A listing of each of the subroutines follows.

```

$IBFTC BEGIN LIST
SUBROUTINE BEGIN ( I )
DIMENSION TABLE(90),TABLE1(135)
COMMON /ONEY/ TABLE, TABLE1

```

```

C
CALL TIME (1)
CALL FIXB (1,XT)
T = XT
RETURN
END

```

```

$IBFTC FIXB LIST
SUBROUTINE FIXB (JKO,TIM7)

```

```

C
TIM = FLOAT(JKO)
TIM1 = TIM * 1.E-5
TIM2 = FLOAT(IFIX(TIM1))*1.E+5
TIM3 = TIM - TIM2
TIM4 = TIM3 * 1.E-3
TIM5 = FLOAT(IFIX(TIM4))*1.0E+3
TIM6 = TIM3 - TIM5
TIM7 = TIM3*1.E-5 + TIM6*1.E-8
RETURN
END

```

```

$IBFTC BARN LIST
FUNCTION BARN ( T )
XT = BARN(1.) + T
WRITE(6,2)T,XT
IF(XT.LT. 0.0)XT = ABS(XT)
2 FORMAT(2(5X,E14.7) )
IF(XT.GT. 1.0)GO TO 1
CALL LOOKUP(XT,XG)
BARN = XG
RETURN
1 XT = XT - 1.0
CALL LOOKUP (XT,XG)
BARN = XG
RETURN
END

```

```

$IBFTC LOOKUP LIST
SUBROUTINE LOOKUP (X,PHINV)

```

```

C
DIMENSION TABLE(90),TABLE1(135)
COMMON /ONEY/TABLE, TABLE1
C
FLAG = 0.0

```



```

FLAG1 = 0.0
IF(X .LT. 0.1) GO TO 2
IF(X .LT. 0.9) GO TO 5
X = 1.0 - X
FLAG1 = 1.0
GO TO 2
5 II = IFIX(X*100.)
I = II - 7
X0 = FLOAT(II) / 100.
X1 = X0 + 0.01
P = (X-X0) / (X1-X0)
4 PS = P*P
PC = PS*P
PF = PC*P
P5 = PF*P
AM2 = 0.83333334E-2 * (P*6.0-PS*5.0-PC*5.0+PF*5.0-P5)
AM1 = -0.41666667E-1 * (P*12.0-PS*16.0+PC*4.0-P5)
A0 = 0.83333334E-1 * (12.0-P*4.0-PS*15.0+PC*5.0+PF*3.0-P5)
A1 = 0.83333334E-1 * (P*12.0+PS*8.0-PC*7.0-PF*2.0+P5)
A2 = -0.41666667E-1 * (P*6.0+PS-PC*7.0-PF+P5)
A3 = 0.83333334E-2 * (P*4.0-PC*5.0+P5)
IF(FLAG1 .NE. 0.0) GO TO 3
PHINV = AM2*TABLE(I-2) + AM1*TABLE(I-1) + A0*TABLE(I)
1 + A1*TABLE(I+1) + A2*TABLE(I+2) + A3*TABLE(I+3)
IF(FLAG1 .EQ. 0.0) GO TO 6
PHINV = -PHINV
X = 1.0 - X
6 RETURN
2 Z = ALOG(X)
IX = IFIX(Z*10.) / 2
X0 = (FLOAT(IX)+2.0 - 2.0) * 0.1
X1 = X0 + 0.2
P = (Z-X0) / (X1-X0)
I = IABS(IX) - 7
FLAG = 1.0
GO TO 4
3 PHINV = AM2*TABLE1(I+2) + AM1*TABLE1(I+1) + A0*TABLE1(I) +
1 A1*TABLE1(I-1) + A2*TABLE1(I-2) + A3*TABLE1(I-3)
IF(FLAG1 .EQ. 0.0) GO TO 7
PHINV = -PHINV
X = 1.0 - X
7 RETURN
END

```

```

$IBFCT BARNALIST
FUNCTION BARNAL(T)
XT = BARN(1.) + T
IF(XT .GT. 1.0) GO TO 1
BARNAL = XT
RETURN
1 BARNAL = XT - 1.0
RETURN
END

```

410010 BLOC LIST

BLOCK DATA

COMMON /ONE/ TABLE, TABLE1

DIMENSION TABLE(100), TABLE1(135)

DATA (TABLE1(1), I=1, 40) /

X -0.10005822E 01, -0.10005822E 01, -0.10005822E 01, -0.10005822E 01,
X -0.10333769E 01, -0.10333769E 01, -0.10333769E 01, -0.10333769E 01,
X -0.07291052E 00, -0.11015135E 01, -0.12222678E 01, -0.13363460E 01,
X -0.21702071E 01, -0.22483373E 01, -0.23244072E 01, -0.23985641E 01,
X -0.24703386E 01, -0.25416478E 01, -0.26107965E 01, -0.26784793E 01,
X -0.27447820E 01, -0.28097820E 01, -0.28735504E 01, -0.29361520E 01,
X -0.29976459E 01, -0.30588858E 01, -0.31175251E 01, -0.31760073E 01,
X -0.32335765E 01, -0.32902727E 01, -0.33461332E 01, -0.34011926E 01,
X -0.34554832E 01, -0.35090356E 01, -0.35618778E 01, -0.36140367E 01,
X -0.36655375E 01, -0.37164033E 01, -0.37665569E 01, -0.38163197E 01 /

DATA (TABLE1(1), I=41, 80) /

X -0.38654090E 01, -0.39139462E 01, -0.39619492E 01, -0.40094320E 01,
X -0.40564132E 01, -0.41029076E 01, -0.41489291E 01, -0.41944917E 01,
X -0.42395085E 01, -0.42842922E 01, -0.43285543E 01, -0.43724068E 01,
X -0.44158600E 01, -0.44589247E 01, -0.45016109E 01, -0.45439291E 01,
X -0.45853856E 01, -0.46274920E 01, -0.46687561E 01, -0.47096856E 01,
X -0.47502837E 01, -0.47905725E 01, -0.48305453E 01, -0.48702127E 01,
X -0.49098823E 01, -0.49486693E 01, -0.49874531E 01, -0.50259663E 01,
X -0.50642063E 01, -0.51021785E 01, -0.51398894E 01, -0.51773410E 01,
X -0.52145419E 01, -0.52514954E 01, -0.52882068E 01, -0.53246804E 01,
X -0.53603209E 01, -0.53969324E 01, -0.54327192E 01, -0.54682855E 01 /

DATA (TABLE1(1), I=81, 120) /

X -0.55036353E 01, -0.55387720E 01, -0.55735998E 01, -0.56084222E 01,
X -0.56429429E 01, -0.56772649E 01, -0.57113916E 01, -0.57453267E 01,
X -0.57792799E 01, -0.58126335E 01, -0.58460113E 01, -0.58792003E 01,
X -0.59122305E 01, -0.59450774E 01, -0.59777527E 01, -0.60102589E 01,
X -0.60425992E 01, -0.60747752E 01, -0.61067899E 01, -0.61386456E 01,
X -0.61703447E 01, -0.62019821E 01, -0.623332813E 01, -0.62645233E 01,
X -0.62956174E 01, -0.63265556E 01, -0.63573699E 01, -0.63880320E 01,
X -0.64185500E 01, -0.64489380E 01, -0.64791955E 01, -0.65092995E 01,
X -0.65392780E 01, -0.65691276E 01, -0.65988477E 01, -0.66284402E 01,
X -0.66579051E 01, -0.66872466E 01, -0.67164648E 01, -0.67455636E 01 /

DATA (TABLE1(1), I=121, 135) /

X -0.67745407E 01, -0.68033974E 01, -0.68321396E 01, -0.68607678E 01,
X -0.68992795E 01, -0.69276753E 01, -0.69559645E 01, -0.69741438E 01,
X -0.70022125E 01, -0.70301756E 01, -0.70580284E 01, -0.70857950E 01,
X -0.71134455E 01, -0.71409942E 01, -0.71684439E 01 /

DATA (TABLE1(1), I=1, 40) /

X -0.10005822E 01, -0.10005822E 01, -0.10005822E 01, -0.10005822E 01,
X -0.10333769E 01, -0.10333769E 01, -0.10333769E 01, -0.10333769E 01,
X -0.07291052E 00, -0.09416525E 00, -0.091536509E 00, -0.07782629E 00,
X -0.04162125E 00, -0.00642124E 00, -0.77219321E 00, -0.73984695E 00,
X -0.70630255E 00, -0.67049974E 00, -0.64334540E 00, -0.61281229E 00,
X -0.58284150E 00, -0.55333471E 00, -0.52440051E 00, -0.49585035E 00,
X -0.46769890E 00, -0.43991316E 00, -0.41246313E 00, -0.38532045E 00,
X -0.35906879E 00, -0.33155335E 00, -0.30549079E 00, -0.27931903E 00,
X -0.25334710E 00, -0.22754477E 00, -0.20189348E 00, -0.17637416E 00,
X -0.15096021E 00, -0.12566134E 00, -0.10043371E 00, -0.75267962E 01 /

DATA (TABLE1(1), I=41, 80) /


```

X -0.50153583E-01, -0.25068907E-01, 0.25068907E-01, 0.12565134E-00,
X 0.50153583E-01, 0.75269862E-01, 0.19043371E-00, 0.12565134E-00,
X 0.15095921E-00, 0.17637416E-00, 0.20189348E-00, 0.22754427E-00,
X 0.25334710E-00, 0.27931913E-00, 0.30548079E-00, 0.33185335E-00,
X 0.35845874E-00, 0.38532045E-00, 0.41246313E-00, 0.43991316E-00,
X 0.46762890E-00, 0.49555935E-00, 0.52440051E-00, 0.55339471E-00,
X 0.58284150E-00, 0.61281299E-00, 0.64334540E-00, 0.67448974E-00,
X 0.70530256E-00, 0.73864685E-00, 0.77219321E-00, 0.80642124E-00,
X 0.84162123E-00, 0.87782629E-00, 0.91536509E-00, 0.95416525E-00,
X 0.99445783E-00, 0.10364334E-01, 0.10803193E-01, 0.11263911E-01 /
D. TA (TABLE(I), I=81, 90) /
X 0.11740867E-01, 0.12265281E-01, 0.12815514E-01, 0.13407549E-01,
X 0.14059715E-01, 0.14757910E-01, 0.15547734E-01, 0.16443536E-01,
X 0.17506859E-01, 0.18907936E-01 /
END

```

```

R13MAP RARV
R26 026 4095
R27 TVI R27V
R28 L20 R28V+12,0
R29 WBY R29V+13,0
R30 LLS 0,0
R31 ALS 4,0
R32 LRS 4,0
R33 STO R29V+12,0
R34 ADJ R29V+12,0
R35 STO R29V+12,0
R36 ARS 4,0
R37 ORA R29V+14,0
R38 FAD R29V+14,0
R39 TRA 1,4
R40 OCT 2312421537,1737,2000000000 0
R27V SXD IR1,1
R28V SXD 124,0
R29V LVA L20,1
R30V CLA R29V+14,0
R31V STO 0,0
R32V TSX R29V,4
R33V FAD 0,0
R34V STO 0,0
R35V ITX R29V+5,1,1
R36V FOP L20+1,0
R37V CLA L20+4,0
R38V LLS 35,0
R39V FSD L20+2,0
R40V LRS 35,0
R41V FMP L20+3,0
R42V LXD 121,1
R43V LXD 124,4
R44V TRA 1,4
L20 HTR 20,0
DEC 20,0,05,15,42103340,0
IR1 HTR 0,0
IR4 HTR 0,0

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REPRODUCIBILITY OF THE ORIGINAL PAGE IS POOR.

C

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END

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